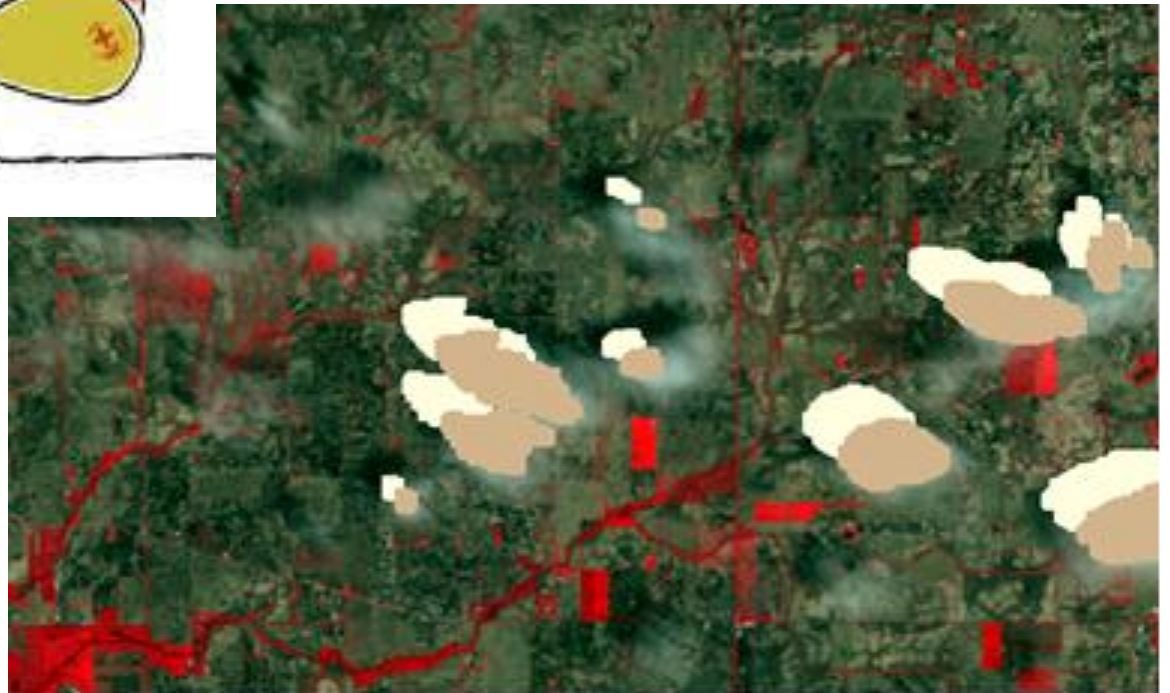


# Making Multitemporal Work

**Randolph H. Wynne, Christine E. Blinn, Kevin J. Boyle, Evan B. Brooks, Harold E. Burkhart, John W. Coulston, Thomas R. Fox, Klaus Moeltner, S. Seth Peery, and Valerie A. Thomas**

Presented to Landsat Science Team  
Sioux Falls, South Dakota  
October 30, 2013

- **Is crowd sourcing a reliable and cost-effective solution to removing clouds and cloud shadows that remain subsequent to the application of automated algorithms?**
- **How can Fourier series best be used to create smooth periodic time series of Landsat data?**
- **Using smooth periodic time series generated from Fourier regression or (as necessary) (E)STARFM (Gao et al. 2006, Zhu et al. 2010), how can both gross and subtle changes to land use / cover be detected reliably using from one to three additional observations?**
- **How well do empirical (Flores et al., 2006) and physically-based (Ganguly et al., 2012) algorithms for leaf area index (LAI) generation estimate LAI in intensively-managed pine ecosystems? How can they be improved?**
- **Using the new tree canopy product as a case study, how can the precision of vegetation continuous field products be estimated in a robust and computationally-efficient manner?**



All HITs | HITs Available To You | HITs Assigned To You

Find **HITs** containing  that pay at least \$  ☐ require Max

**Total Earned: \$0.00**  
**Total HITs Submitted: 1**

Skip HIT

 **Your results have been submitted to Virginia Tech Cloud Interpretation Group and will be approved or rejected shortly.**

You can work on this new HIT by clicking the "Accept HIT" button.

## Identify clouds on satellite images

**Requester:** Virginia Tech Cloud Interpretation Group

**Reward:** \$0.50 per HIT

**HITs Available: 6**

**Duration:** 8 hours

**Qualifications Required:** None

**You are previewing this HIT.**

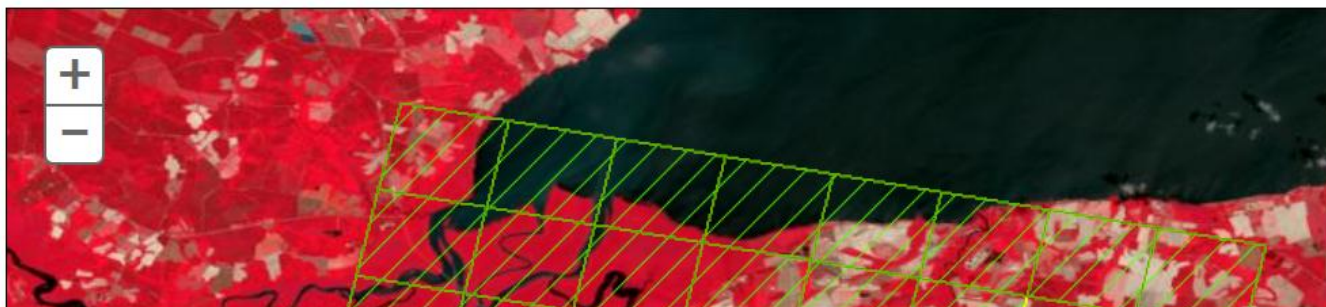
- You must provide informed consent to participate in this research.
- **You can only complete one HIT in this phase of our research study.**
- JavaScript must be enabled in your browser for this HIT to work properly.
- If you receive a "mixed content" warning, see the mTurk help article [here](#).

### 1) Informed Consent

## 2) Training Module

### 3) Cloud Interpretation Tasks

#### 4) Exit Survey





**amazon**mechanical turk  
Artificial Intelligence

Virginia Tech Cloud Interpretation Group | [Account Settings](#) | [Sign Out](#)

Your Account | **HITS** | Qualifications | 409,372 HITS available now

All HITS | HITS Available To You | HITS Assigned To You

Find **HITS** containing [ ] that pay at least \$ 0.00 ☐ for which you are qualified ☐ require Master Qualification **GO**

Timer: 00:00:59 of 8 hours

Finished with this HIT?  Let someone else do it?

☐ Automatically accept the next HIT

Total Earned: \$0.00  
Total HITS Submitted: 0

Identify clouds on satellite images

Requester: Virginia Tech Cloud Interpretation Group

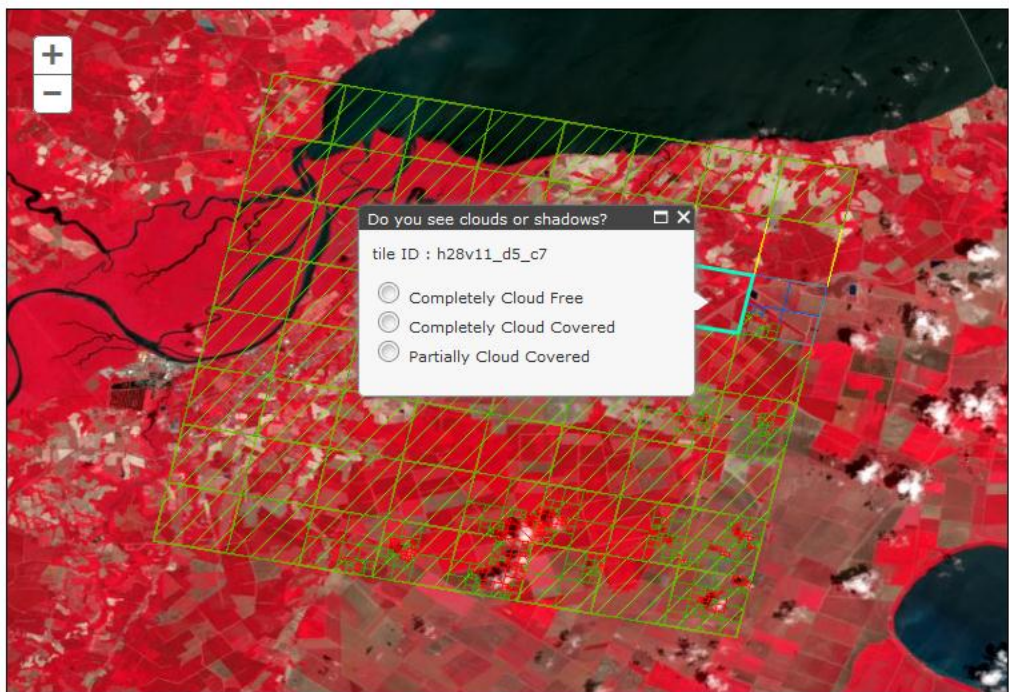
Qualifications Required: None

Reward: \$0.50 per HIT | HITS Available: 6 | Duration: 8 hours

Your Assignment ID is: 2G2UC26DG67J7BDCHKB0FOK2K7ZD6X

- You must provide informed consent to participate in this research.
- **You can only complete one HIT in this phase of our research study.**
- JavaScript must be enabled in your browser for this HIT to work properly.
- If you receive a "mixed content" warning, see the mTurk help article [here](#).

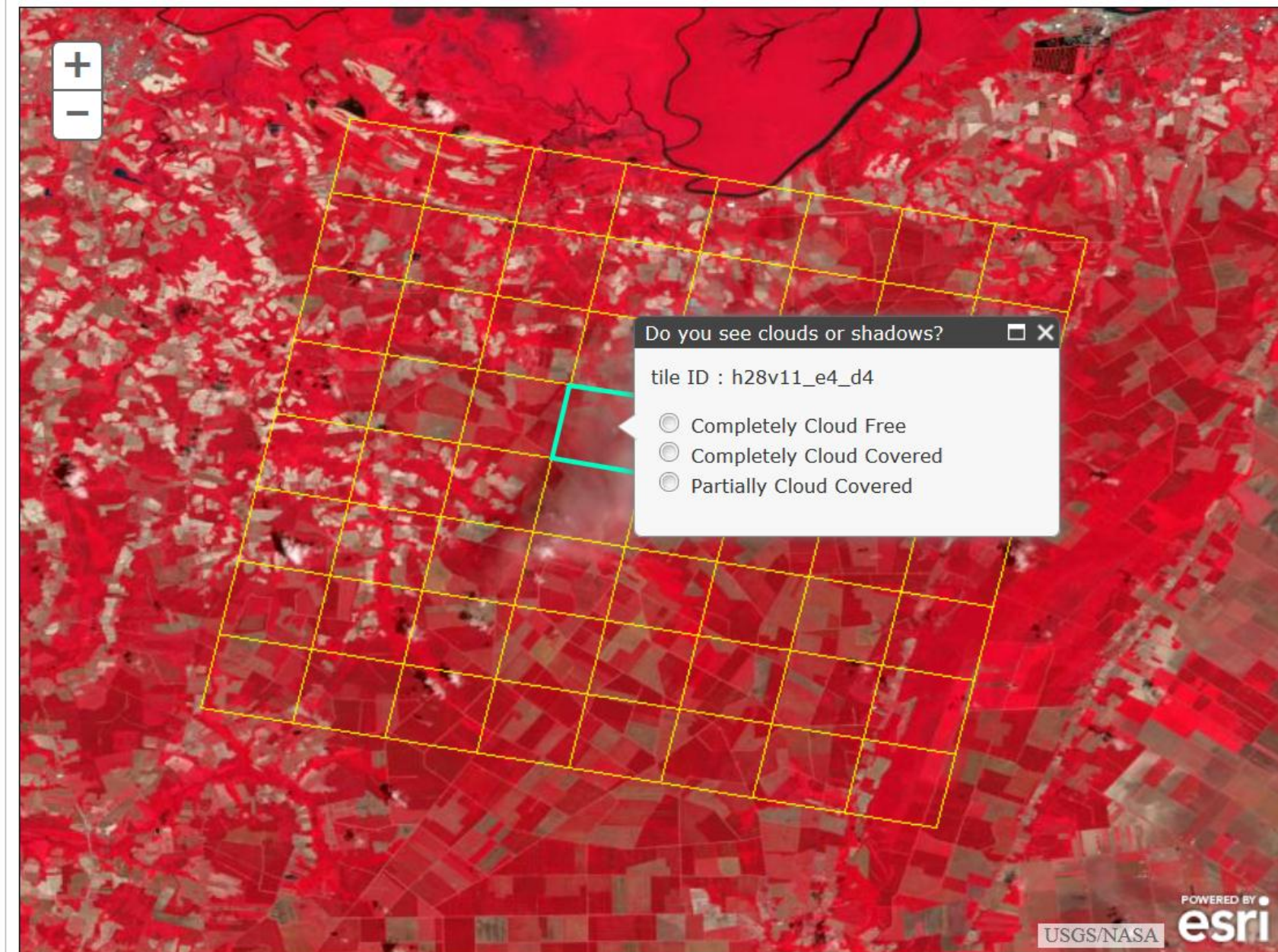
- 1) Informed Consent
- 2) Training Module
- 3) Cloud Interpretation Tasks
- 4) Exit Survey





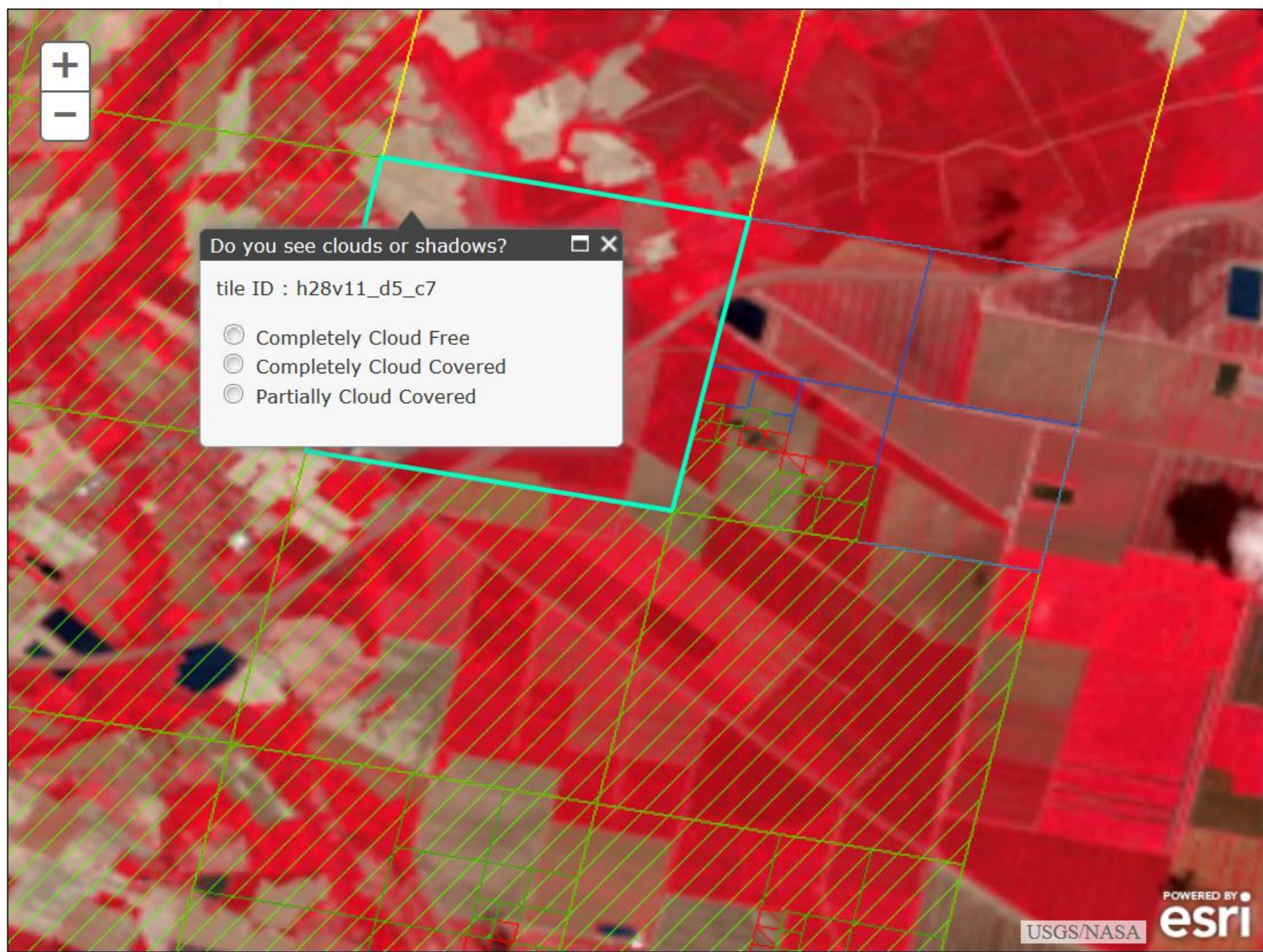
Task 3 of 3: Find clouds in this scene. When you have finished the task, press the Continue button.

1) Informed Consent 2) Training Module 3) Cloud Interpretation Tasks 4) Exit Survey

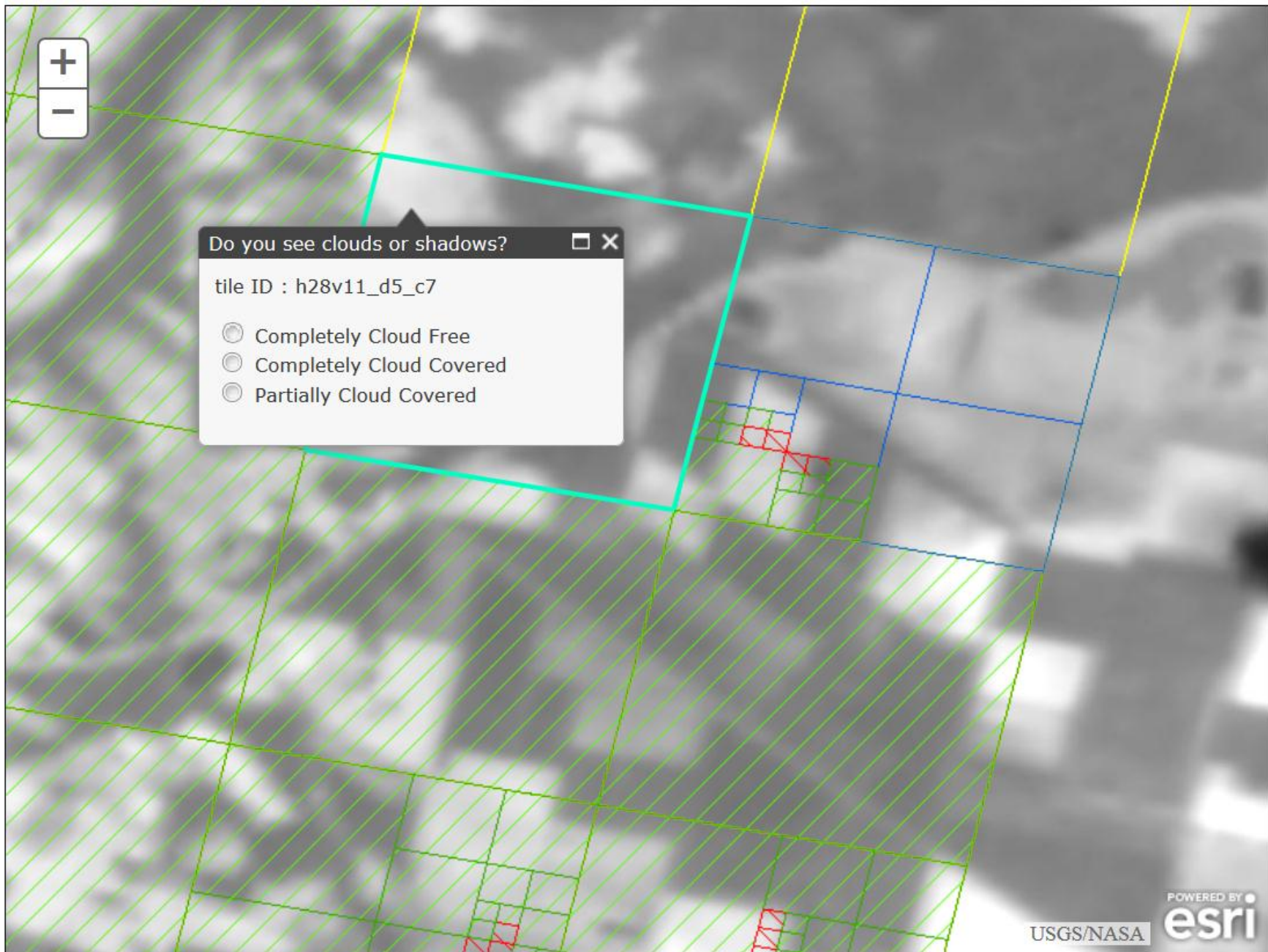


Change image display mode Color infrared Continue









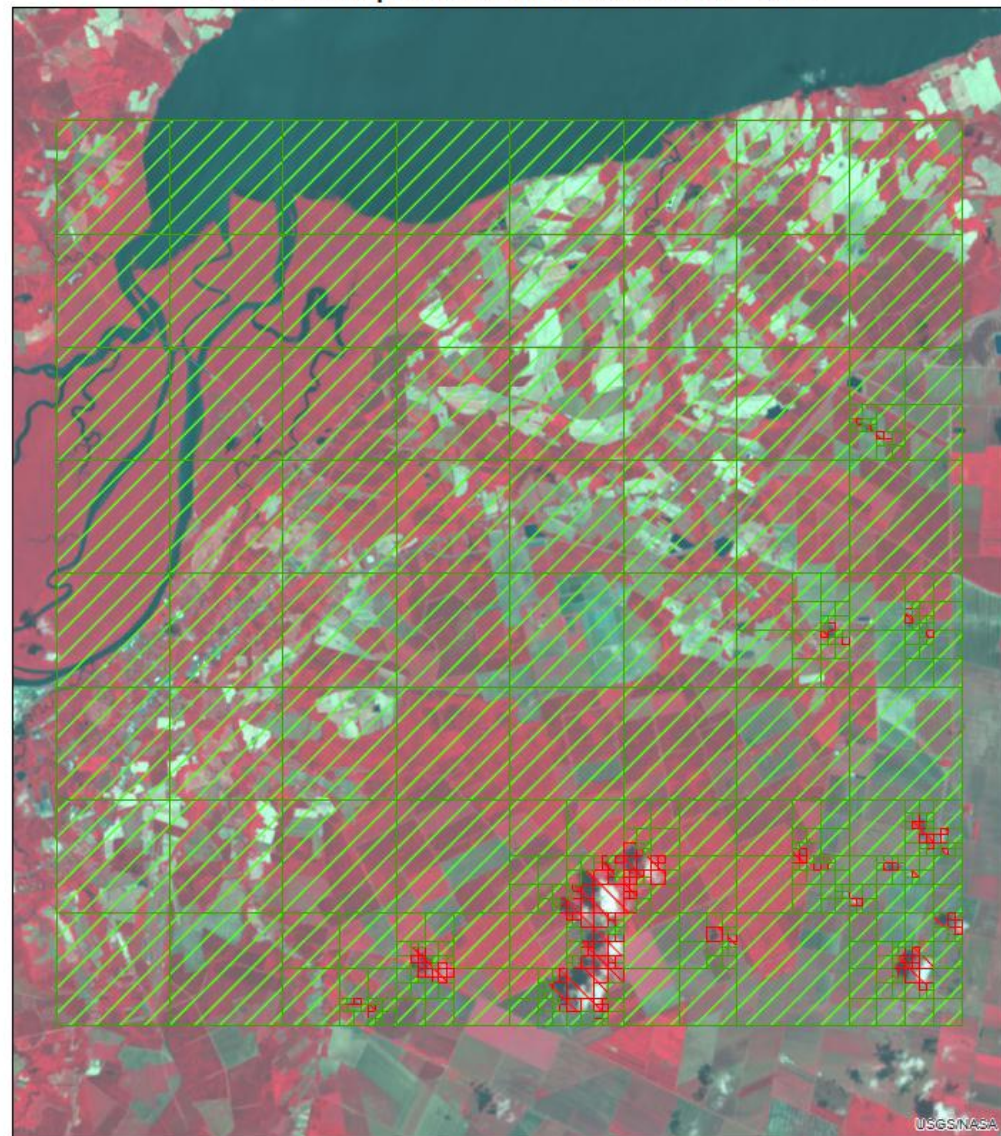






A sample scene



0 1.25 2.5 5 7.5 10 Kilometers

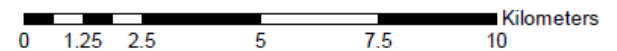
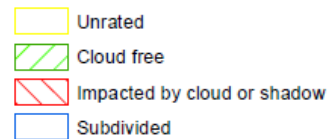
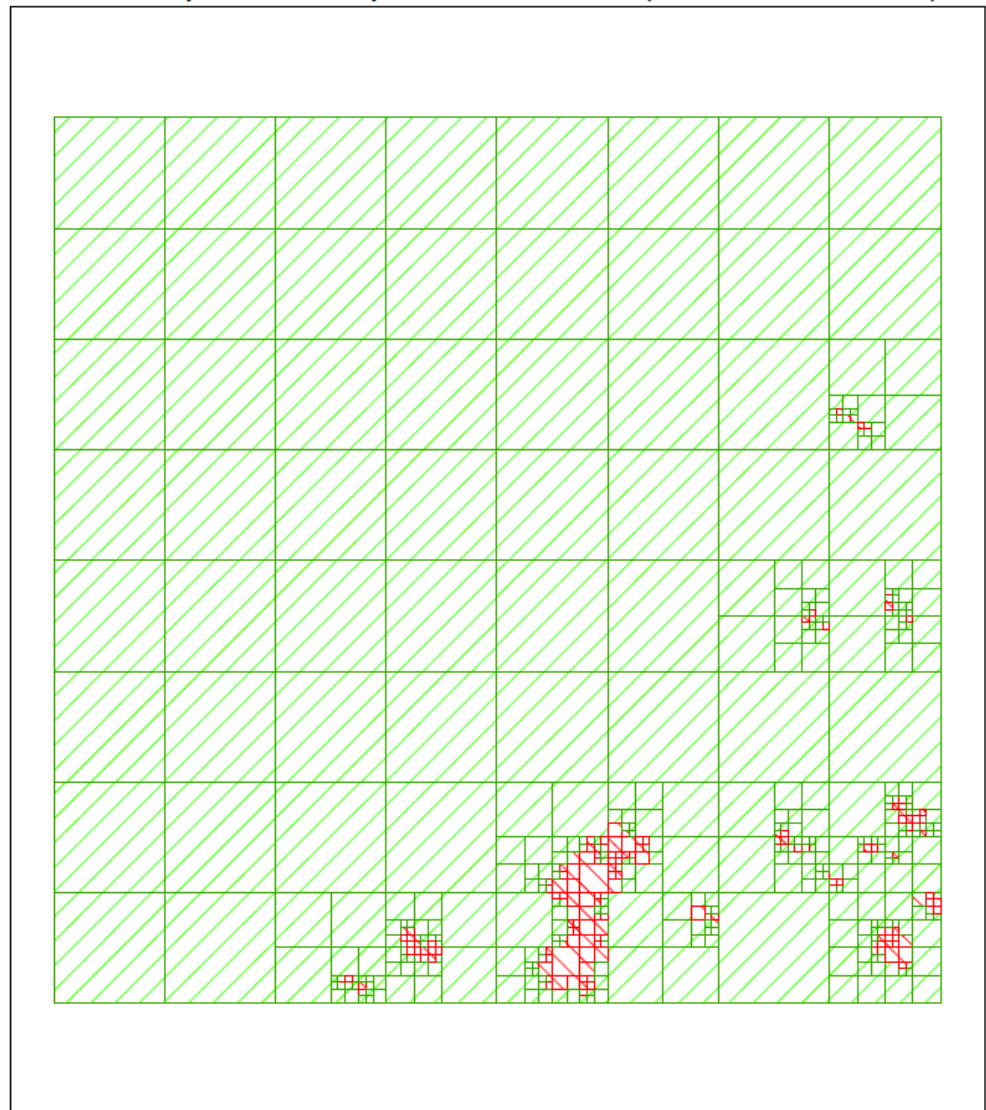
## Full Interpretation Task over CIR



-  Unrated
-  Cloud free
-  Impacted by cloud or shadow
-  Subdivided

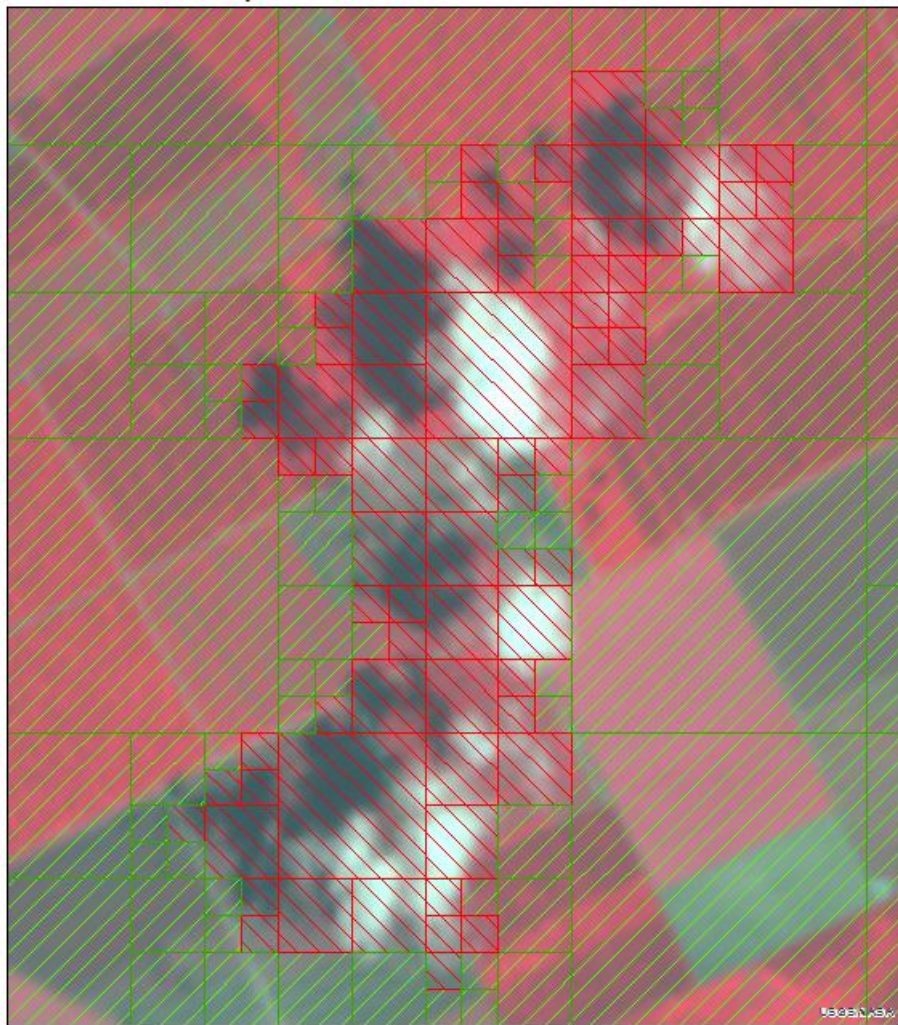
0 1.25 2.5 5 7.5 10 Kilometers







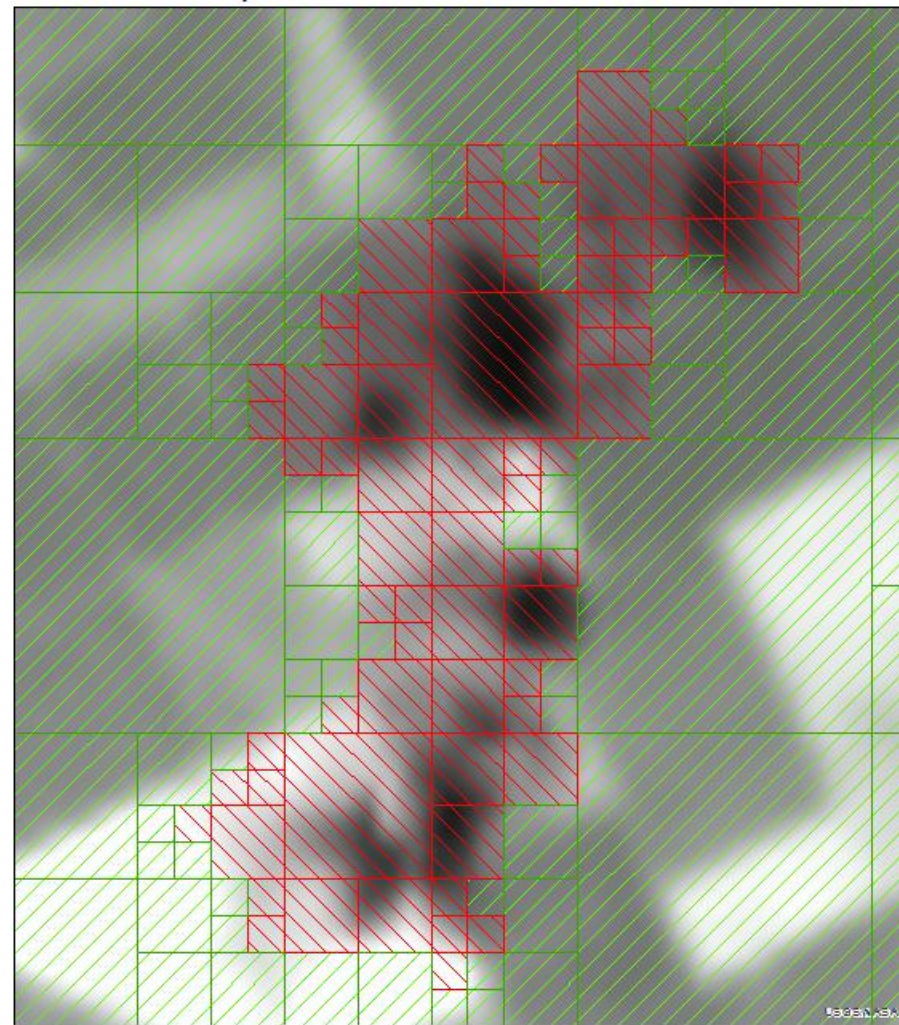
Interpretation detail view over CIR



- Unrated
- Cloud free
- Impacted by cloud or shadow
- Subdivided

0 0.2 0.4 0.8 1.2 1.6 Kilometers

Interpretation detail view over TIRS

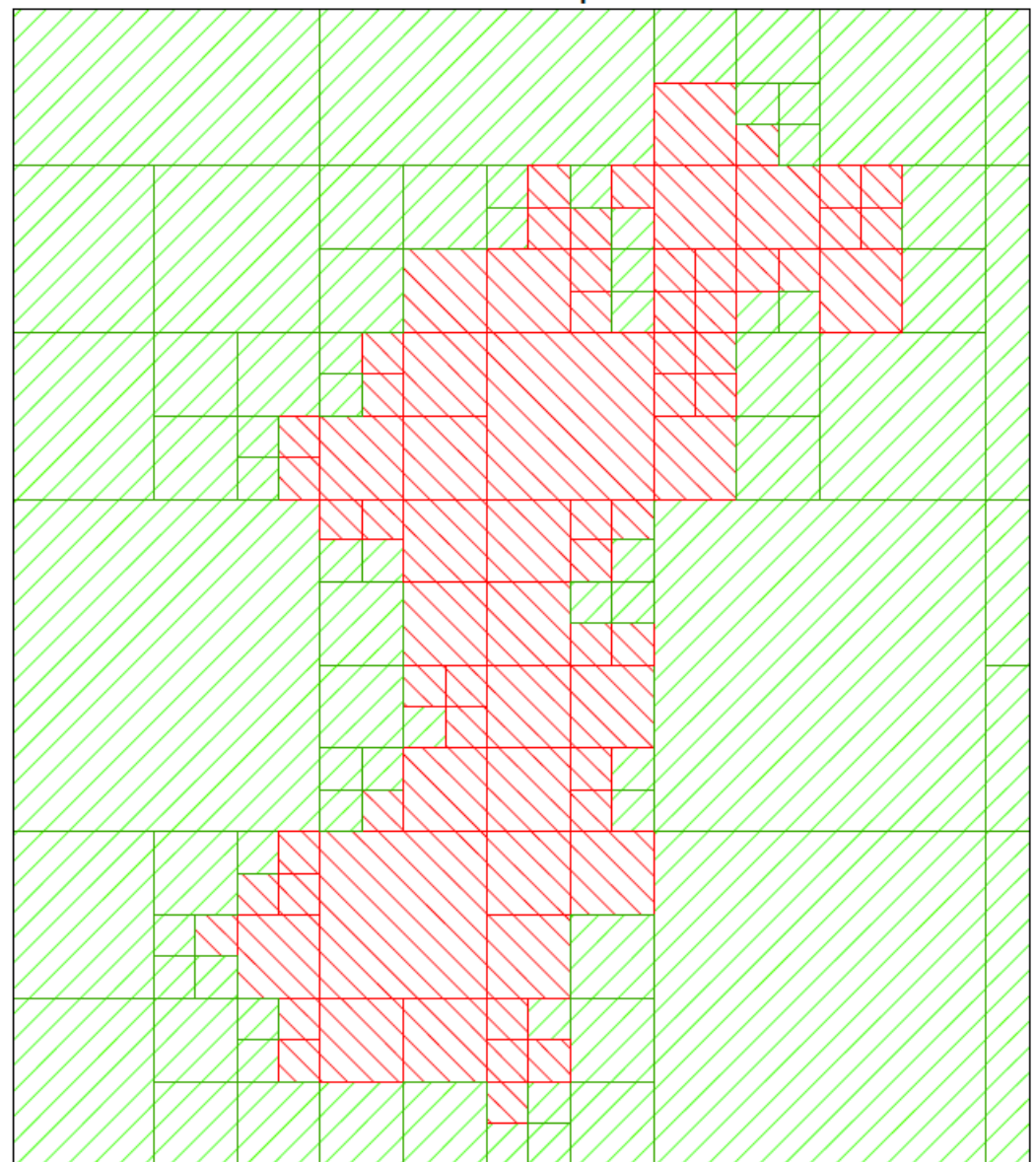






- Unrated
- Cloud free
- Impacted by cloud or shadow
- Subdivided


0 0.2 0.4 0.8 1.2 1.6 Kilometers



## A cloud as a quadtree



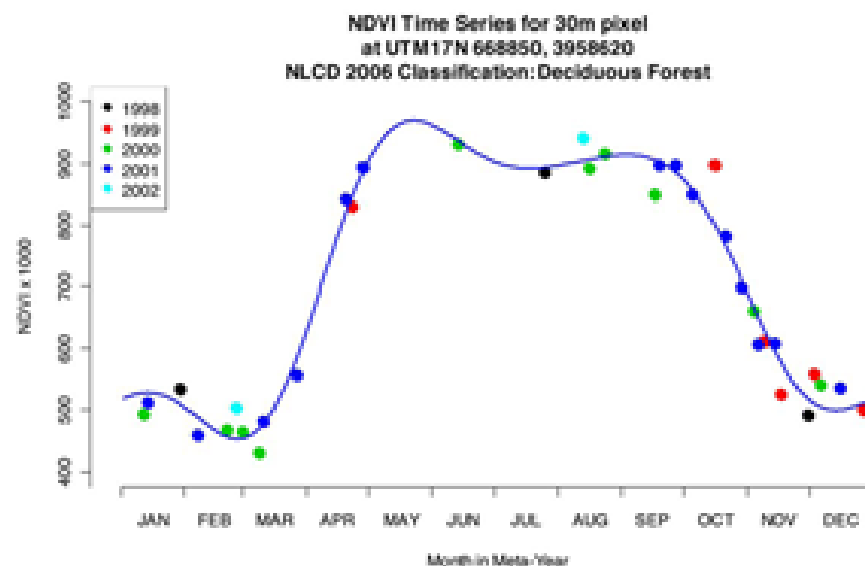
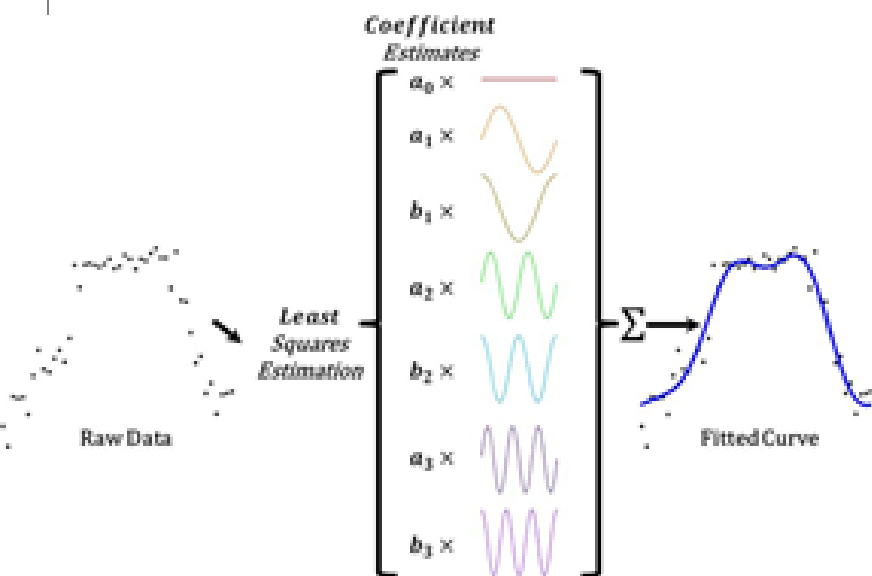
-  Unrated
-  Cloud free
-  Impacted by cloud or shadow
-  Subdivided

 Kilometers  
0 0.2 0.4 0.8 1.2 1.6

# Talking Points: Architecture

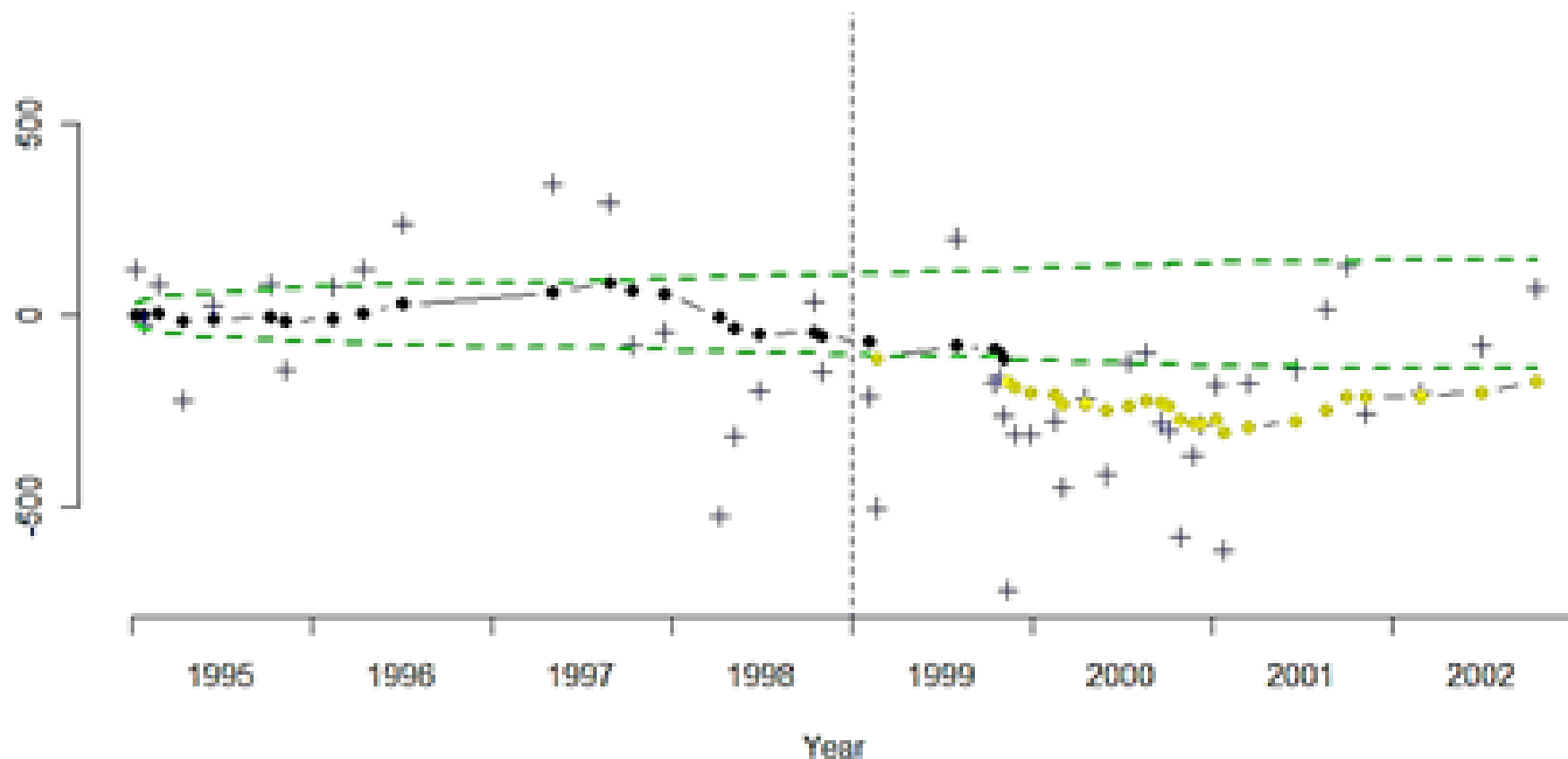
- **The infrastructure that we have developed for the current use case of identifying clouds on Landsat scenes has the potential to serve as a general purpose platform for the integration of geospatial crowdsourcing tasks into mTurk.**
- **The quadtree implementation we use allows for interpretation at varying scales.**
- **mTurk provides a workforce. Our system provides a means to present geospatial problems, hosted in ArcGIS Server, to that workforce as External Questions.**





**EWMA Chart for Residual Time Series for 30m Landsat Pixel at  
UTM 16N 458970, 3665580**

Deviation From Curve, Angle Index x1000





# Projects

## ➤ Current Activity

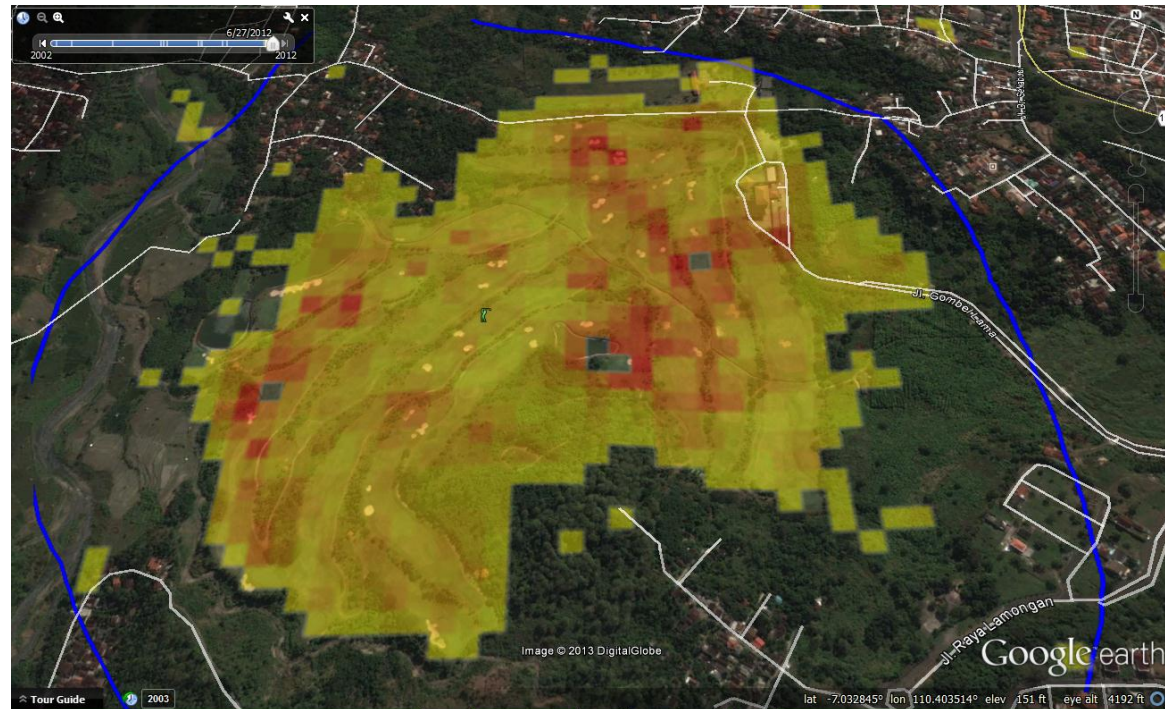
- Exponentially Weighted Moving Average Change Detection (EWMACD) in Java, Indonesia
  - Collaboration with Jeffrey Cardille, Department of Geography, Université de Montréal, Montreal, Quebec, Canada
- Forest Inventory and Analysis (FIA) in Carolinas, USA
  - Collaboration with John Coulston, USDA Forest Service FIA Program, Southern Research Station
  - Change in vegetative cover/dynamic parameter estimation (via EWMACD)
  - Forest biophysical parameter estimation (via harmonic regression)

## ➤ On Deck

- Harmonic Suitability Index (HSI)
- Iterative Hierarchical Cluster Analysis (iHCA)

# Tropical EWMACD

- Focus on 2002-2013 to better match Google Earth imagery for rough validation
- Variety of interesting change features detected

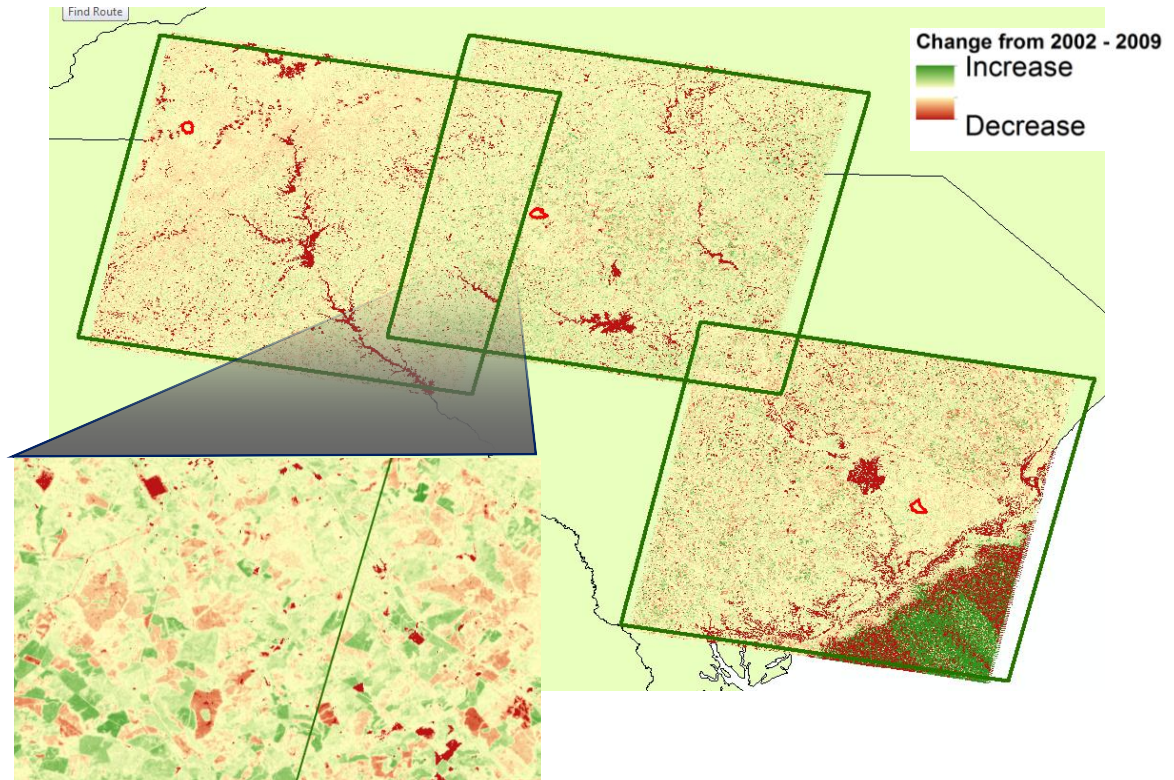




# FIA and Landsat – Vegetation Cover Change

## ➤ Results

- Every Landsat image in 2010 and 2011
  - LEDAPS-processed (Masek et al., 2006)
  - Converted to NDVI
  - Fmask filtering (Zhu and Woodcock, 2012)
  - 2-harmonic curves (Brooks et al., 2012)
- Comparison to FIA plot observations is pending...



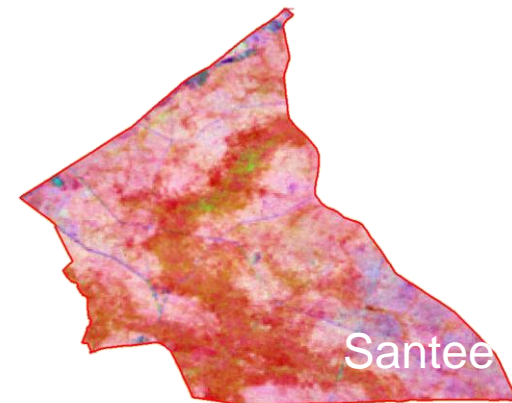
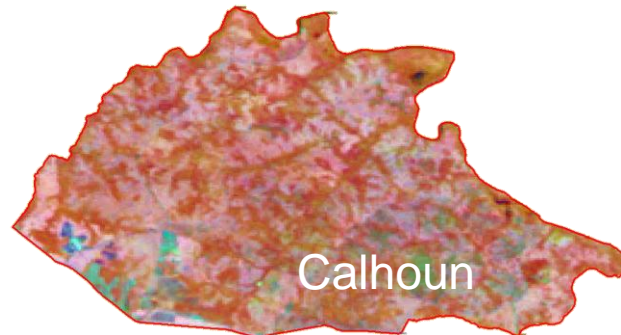
# FIA and Landsat

## ➤ Results

- Coefficients do a reasonably good job of predicting most carbon-related measurements
  - Carbon above ground,
  - Carbon below ground,
  - Carbon in litter, dry biomass in the crown
- Other associations include stand age, stand species type, and type of recent disturbance



R,G,B  
Constant, Sine 1, Cosine 1

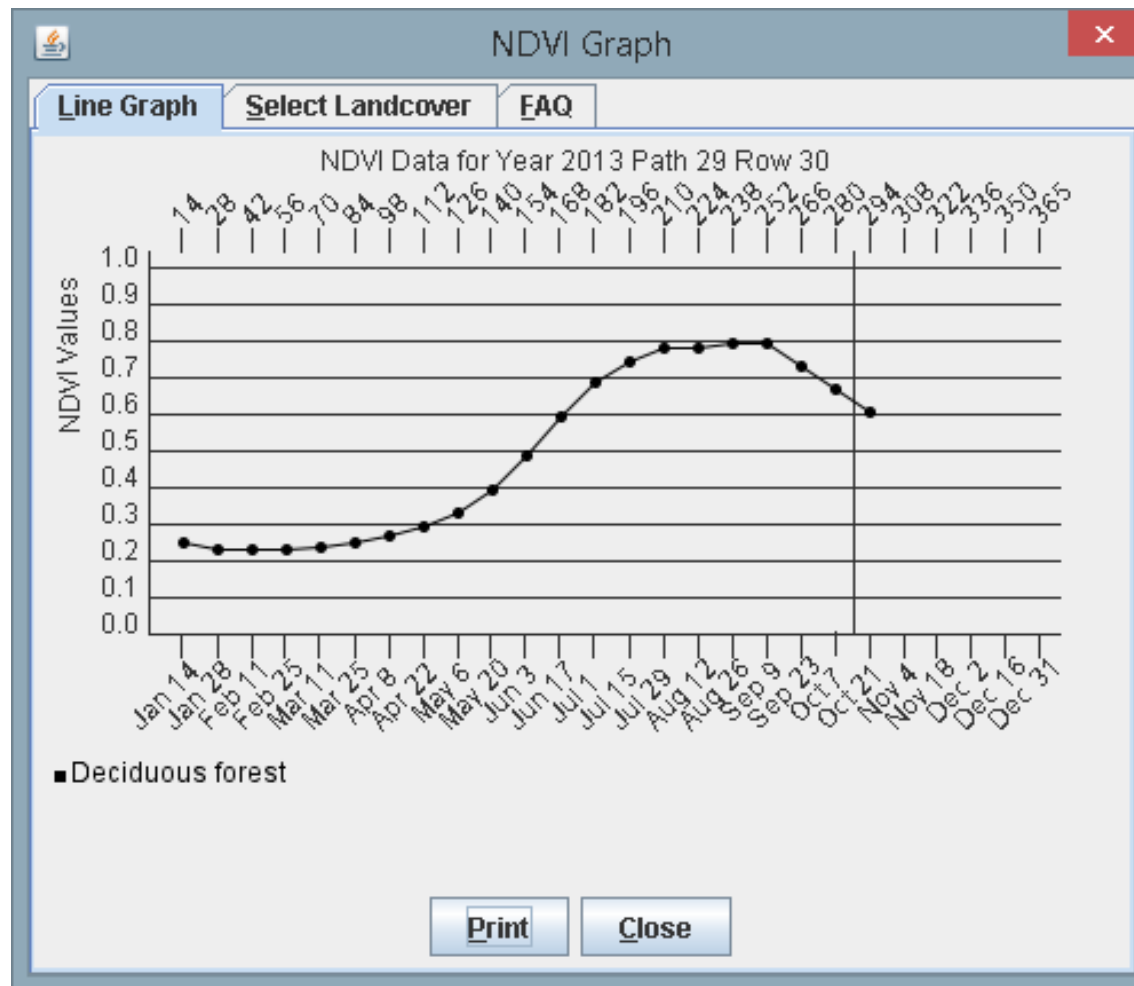




# Harmonic Suitability Index

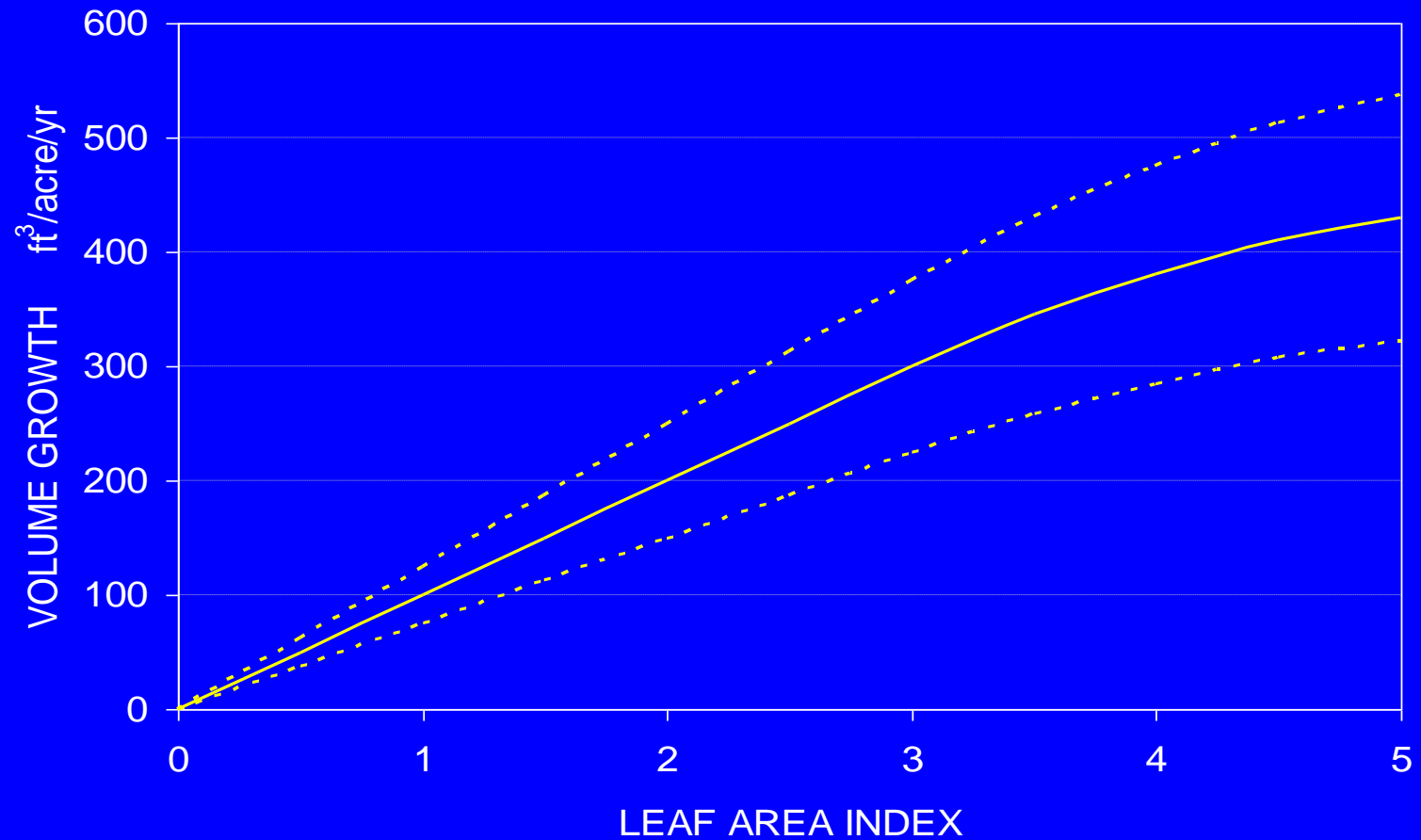
- **Driving Idea:** A single-number statistic, computed solely from temporal information of a stack of images, that gives a sense of how appropriately a harmonic curve will be able to fit that stack
- **Applications:** Virtually anywhere harmonic regression is used
  - Retraining after disturbance signals, automated computing from incoming data

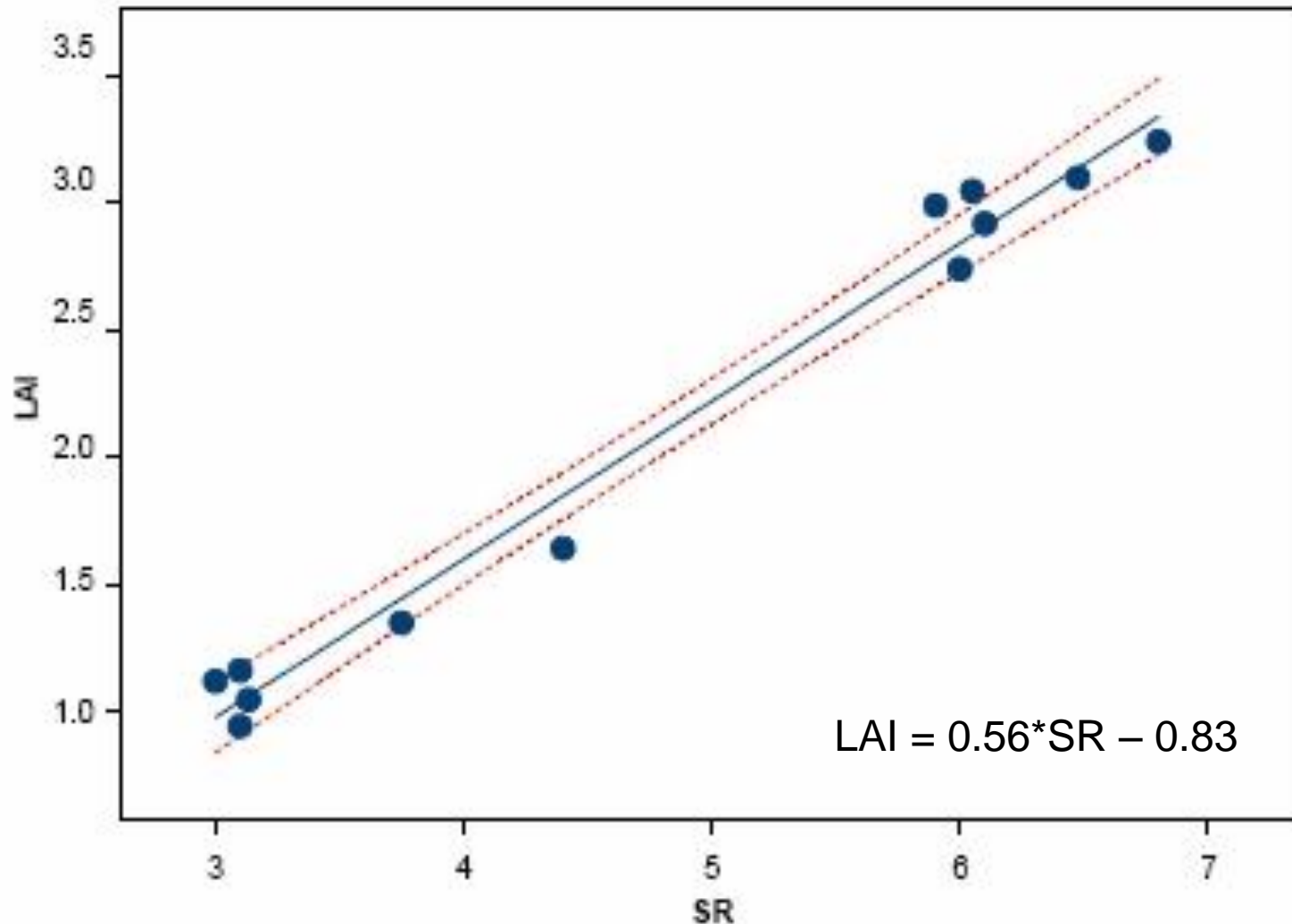
# And if it is NOT suitable?





# LAI is Related to Tree Growth

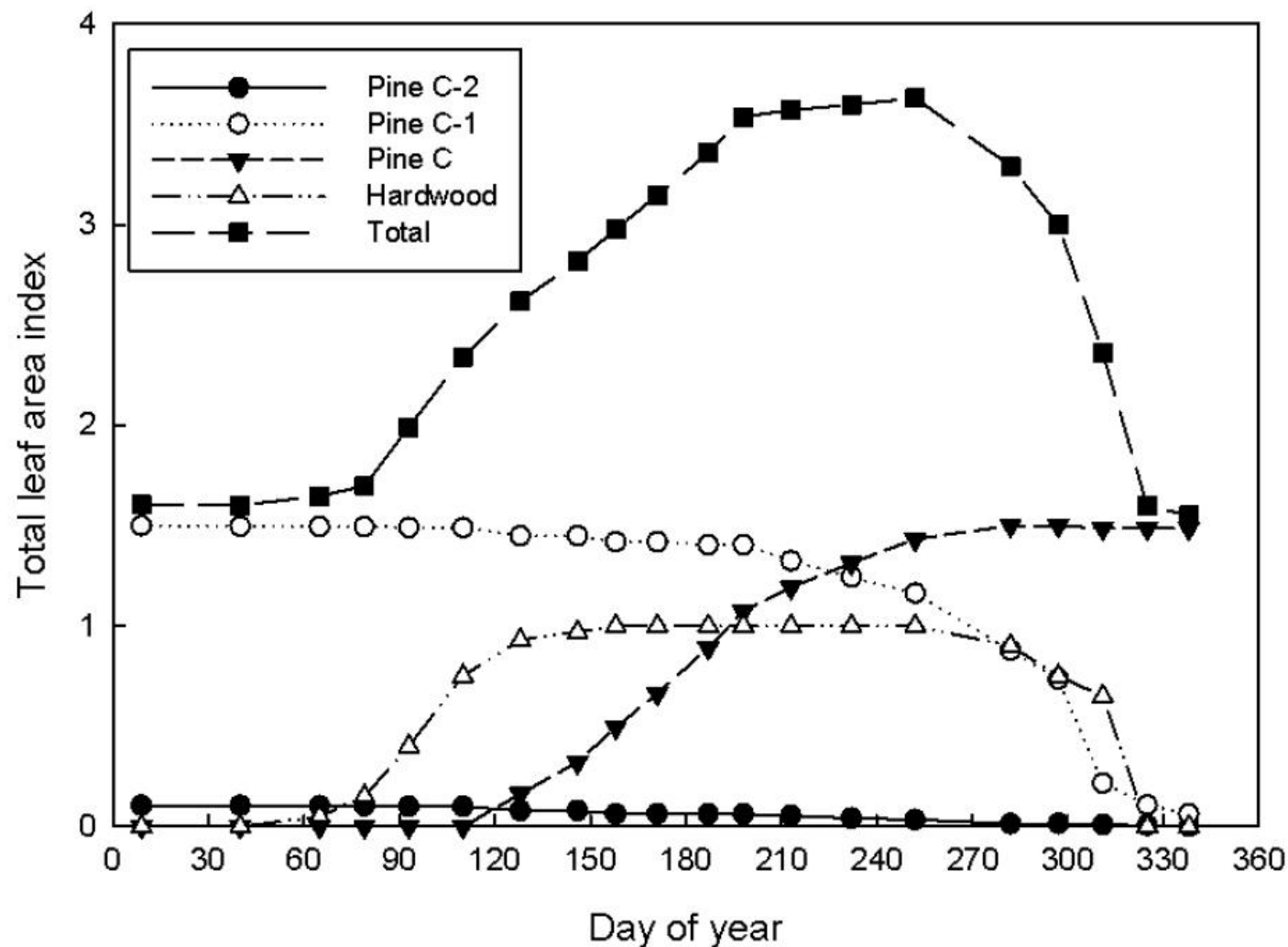




SR = simple ratio vegetation index (Landsat 7)



# Loblolly Pine LAI Dynamics



Where:

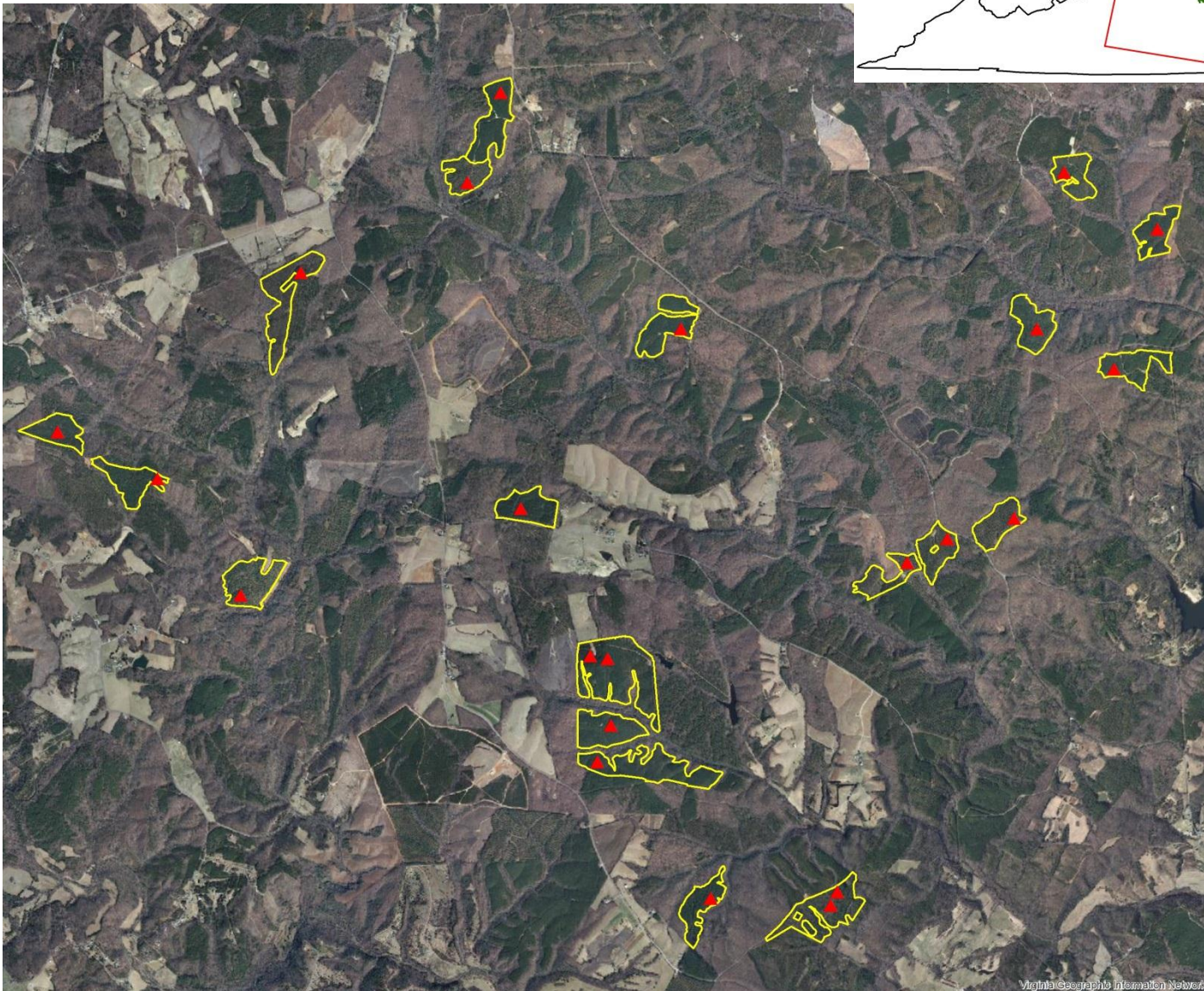
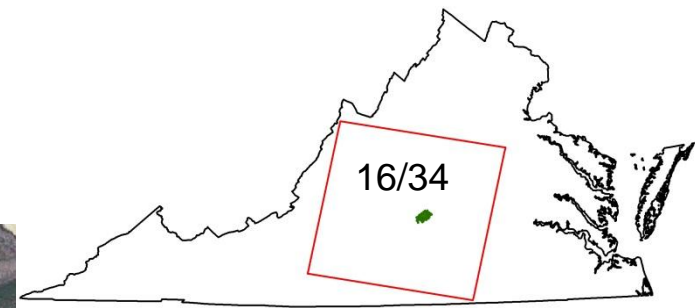
**Pine C-2** = 2-year old needle cohort

**Pine C-1** = 1-year old needle cohort

**Pine C** = current year needle cohort

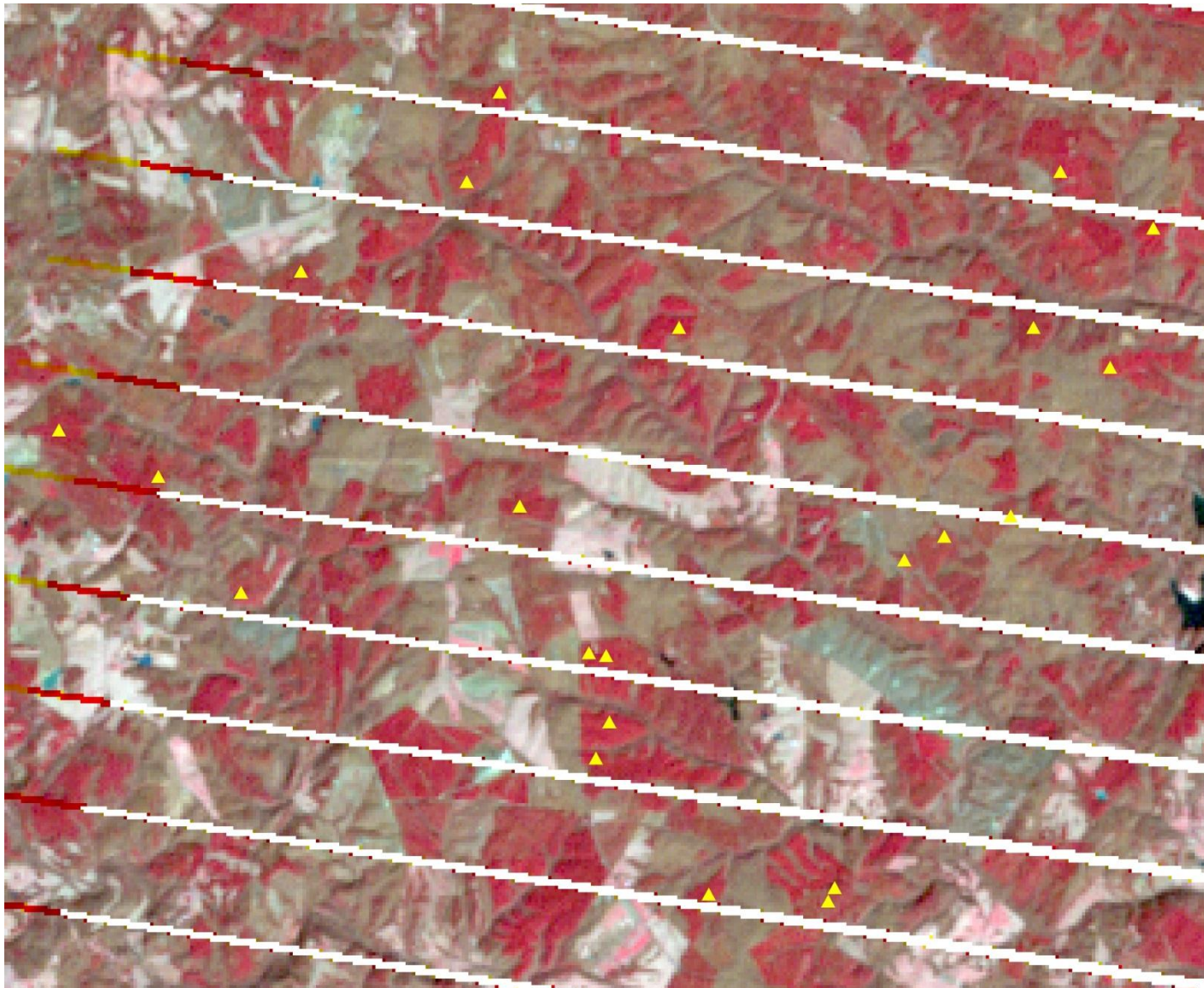
**Hardwood** = hardwood LAI

**Total** = total LAI



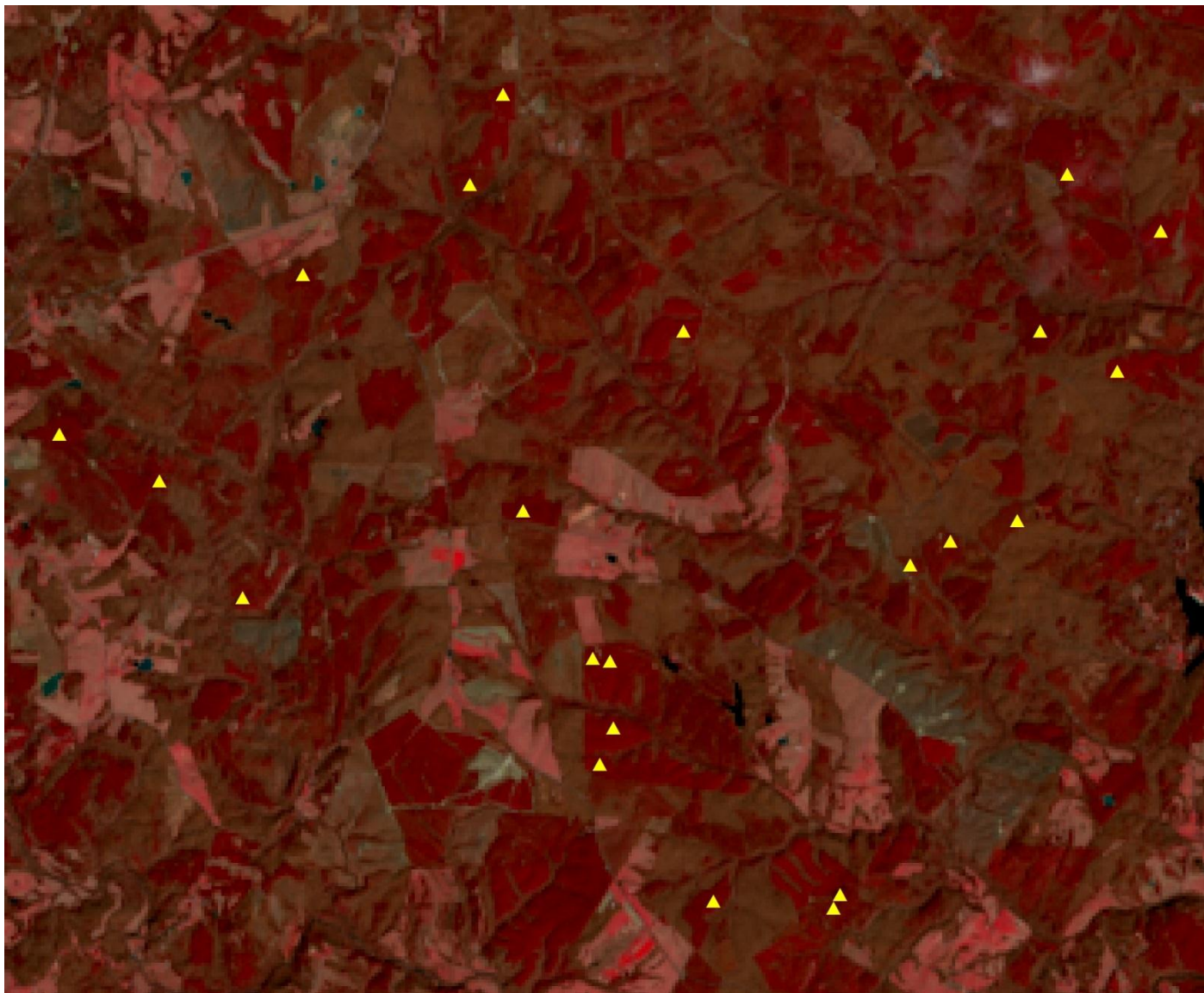


# Landsat 7 w/Plots

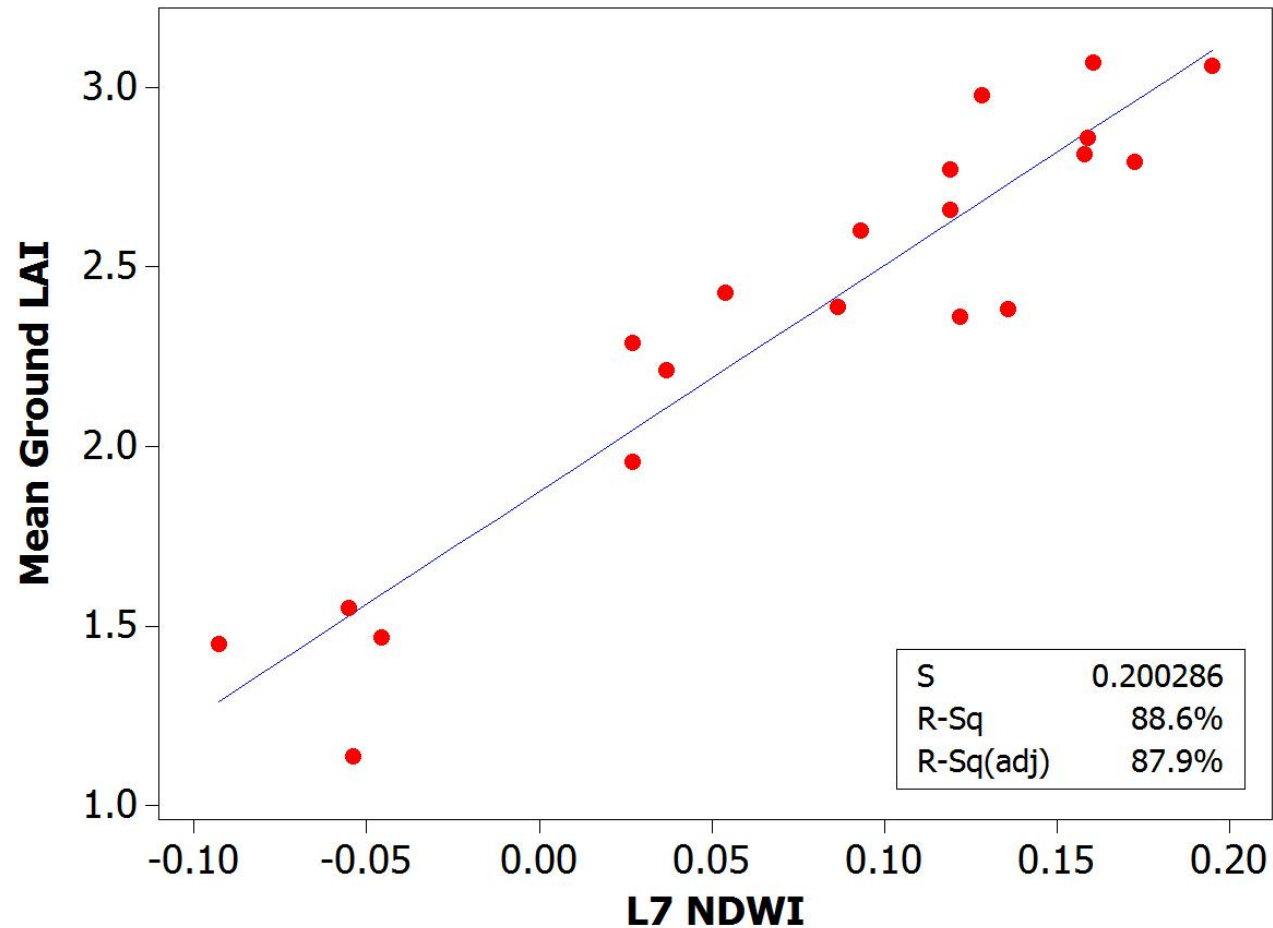




# Landsat 8 w/Plots

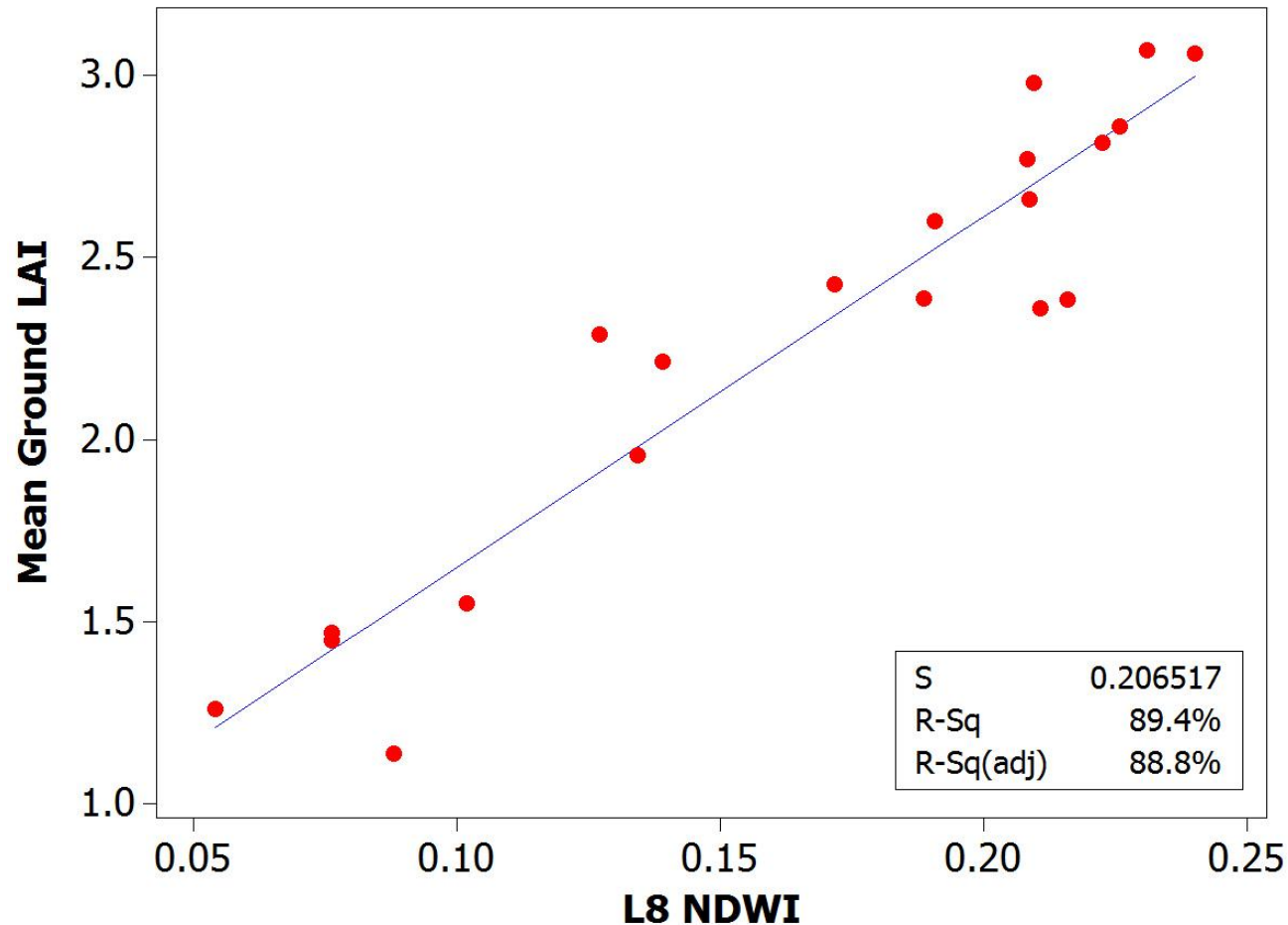


# April Results L7



Note: only 19 of the 20 plots are the same for both L7 and L8

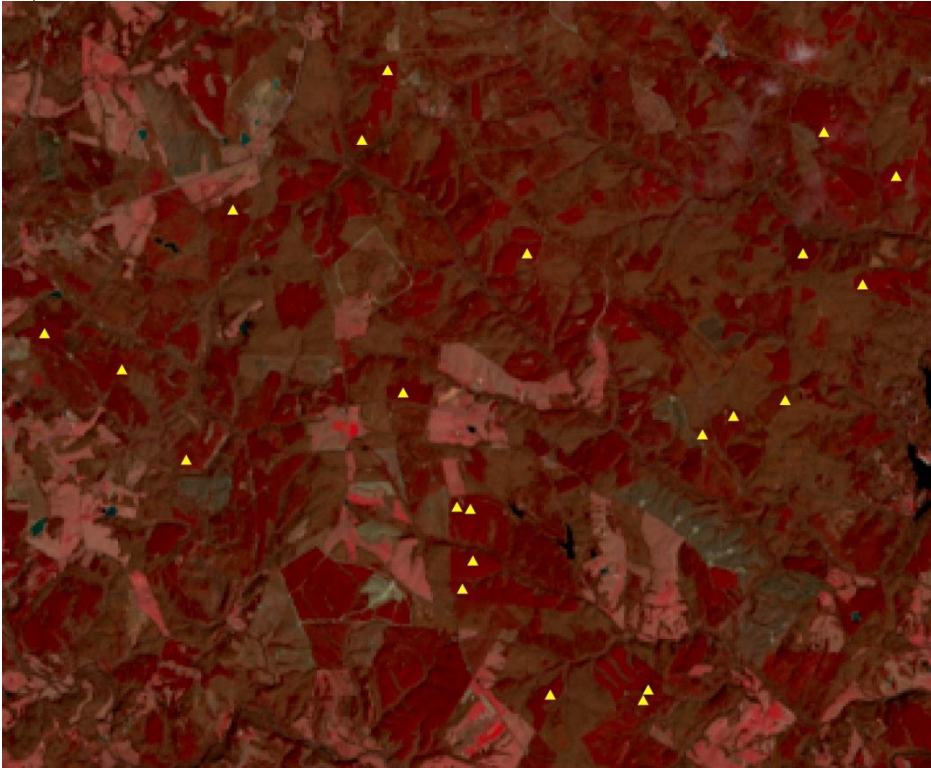
# April Results L8



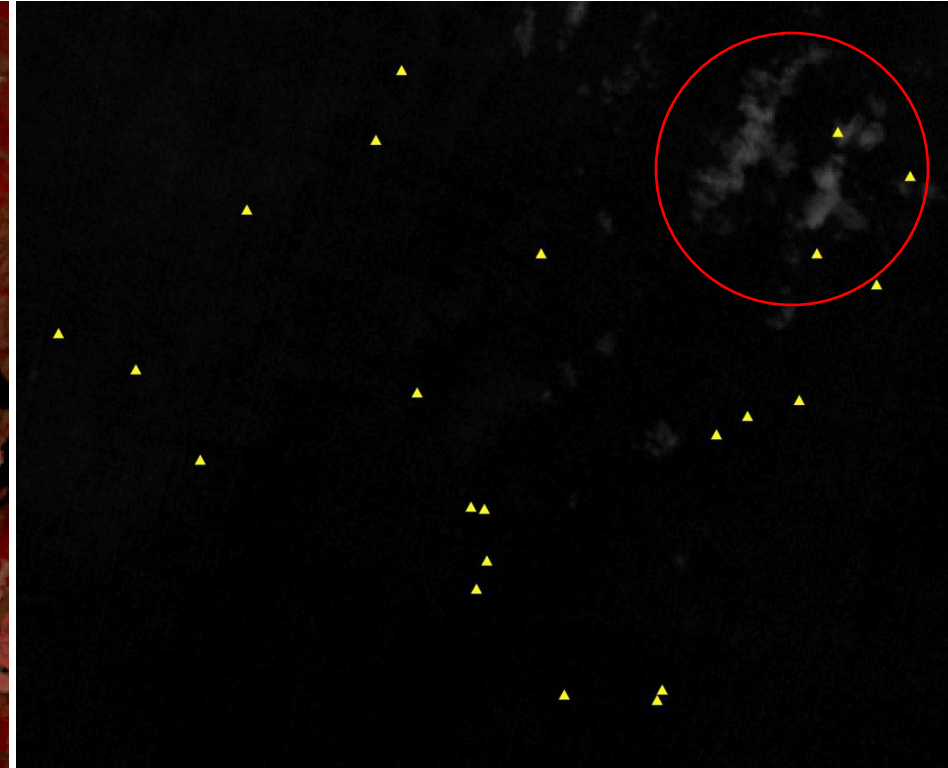
Note: only 19 of the 20 plots are the same for both L7 and L8



# Landsat 8 Cirrus Band



Color-infrared band combination

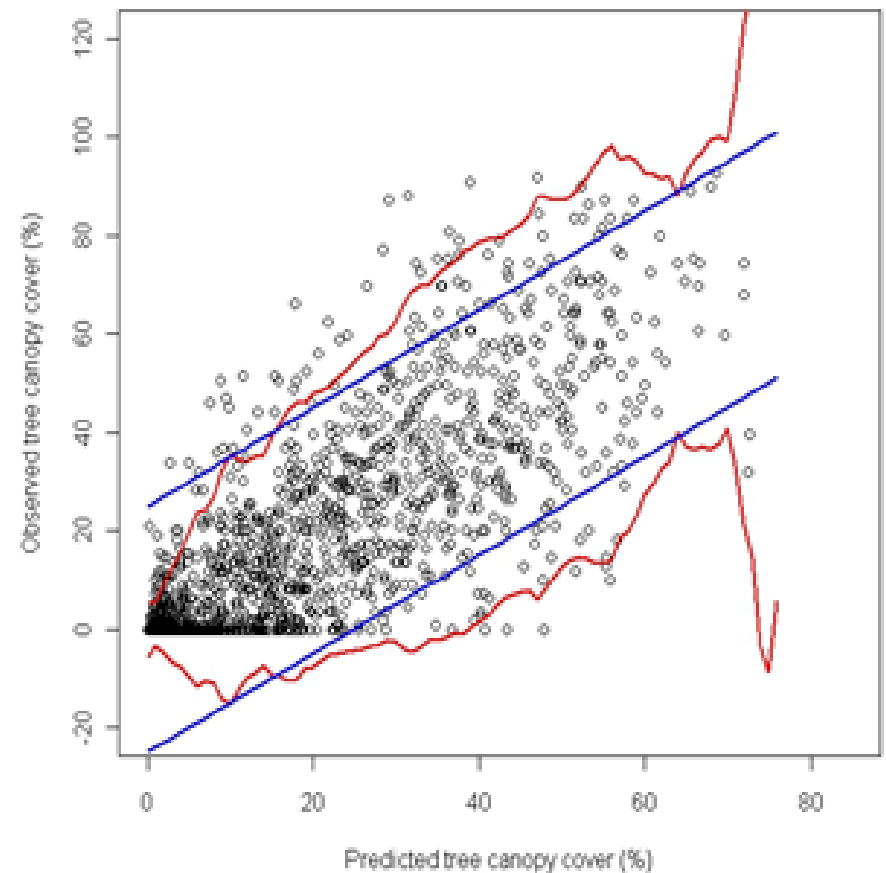


Cirrus Band

March 28, 2013

# **Accomplishments/Findings**

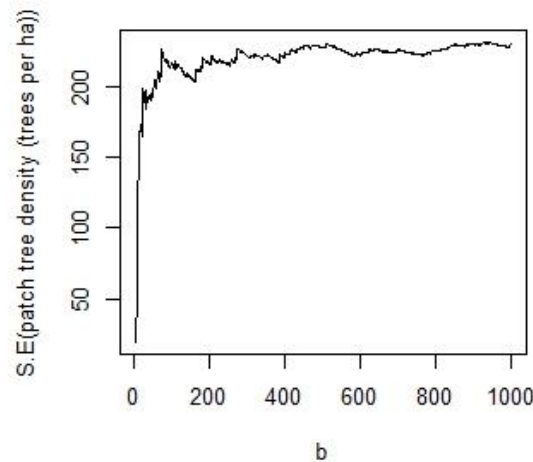
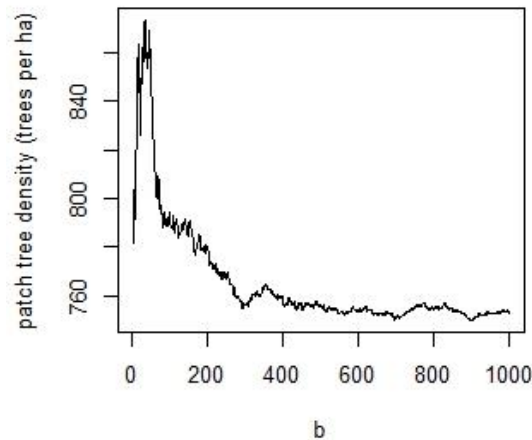
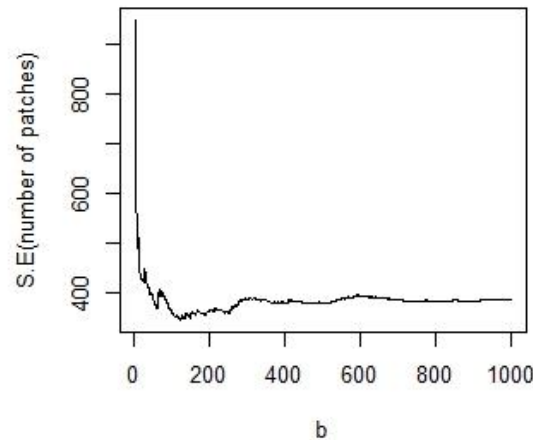
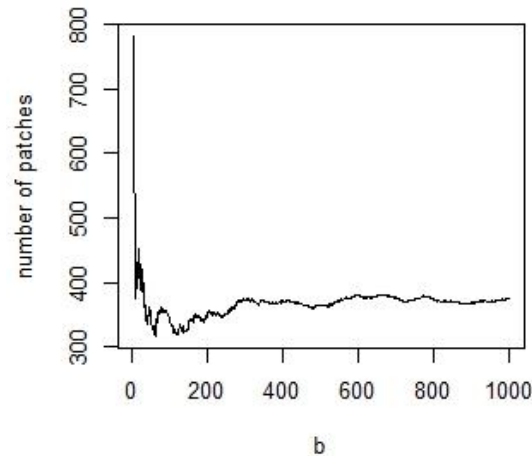
- **Installed a set of permanent plots at one study site for monitoring LAI**
- **NDWI might work better than other VIs when water availability is high**
- **Image timing less critical at minimum LAI, but very important at peak LAI**
- **L8 super tool for empirical LAI estimation**

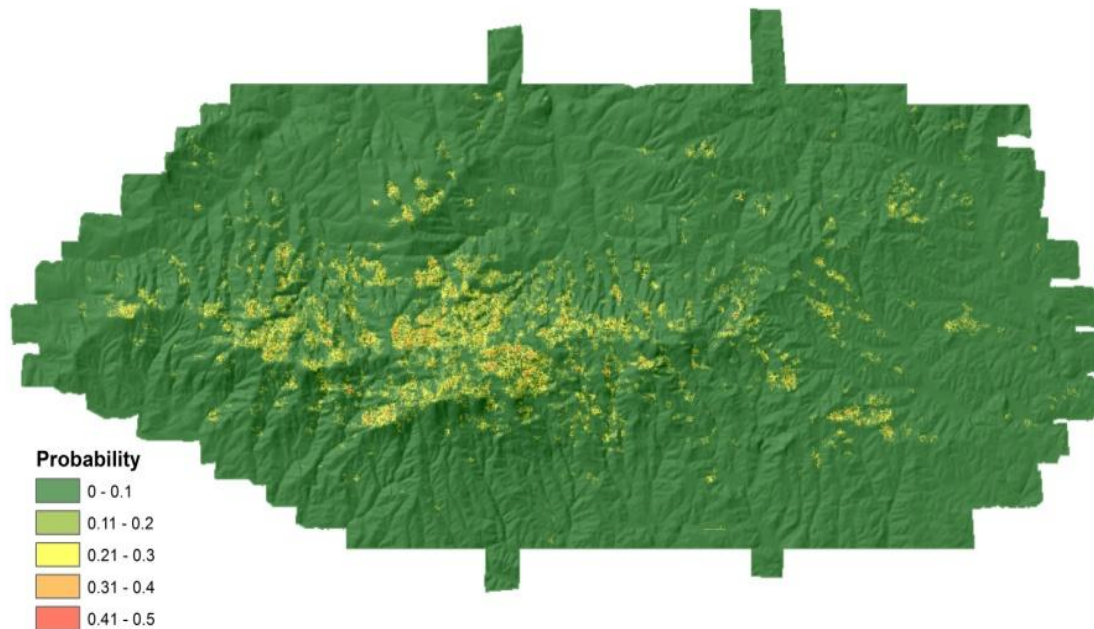


**Figure 10** Example of Monte-Carlo 95% confidence intervals for a random forest model (red) as compared to standard bootstrap 95% confidence interval (blue) for percent tree canopy cover in a sparsely vegetated area of Utah. Note the standard bootstrap confidence intervals are too narrow in parts of the distribution and too wide in other parts of the distribution.



Mean and  
 standard error  
 of the mean  
 for the  
 number of  
 patches and  
 patch tree  
 density based  
 on  $b$   
 bootstrap  
 samples





**Likelihood  
of each  
pixel being  
part of a 1  
ha patch  
with at least  
5 species  
present,  
Moscow  
Mountain**

# Summary I

- **Crowd sourcing a potential solution to removing remnant clouds and cloud shadows using L8 data – but has myriad science applications**
- **Smooth periodic time series generated from Fourier regression enable**
  - Substantial data reduction
  - On-the-fly change detection using control charts
  - Use of harmonic coefficients for continuous and categorical data analysis
  - Need to know where it can't be done and then what



## Summary II

- **Both minimum and maximum LAI well predicted by empirical models using Landsat 8, though NDWI surprisingly best index**
- **The precision of vegetation continuous field products can be estimated in a robust and computationally-efficient manner using Monte Carlo or bootstrapping methods (kNN/MSN or Random Forests)**